

Symmetric Matrices

Defn: A matrix M is symmetric when $M^T = M$.

NB: Because the transpose of an $m \times n$ matrix is $n \times m$, the symmetry condition $M^T = M$ implies M is square.

Ex: $\begin{bmatrix} 0 & 1 \\ 1 & 3 \end{bmatrix}^T = \begin{bmatrix} 0 & 1 \\ 1 & 3 \end{bmatrix}$ is symmetric.

$\begin{bmatrix} 3 & -5 \\ -1 & 0 \end{bmatrix}^T = \begin{bmatrix} 3 & -1 \\ -5 & 0 \end{bmatrix}$ is NOT symmetric.

Ex: The 2×2 real symmetric matrices are:

$$\text{Sym}_2(\mathbb{R}) = \left\{ \begin{bmatrix} a & b \\ b & c \end{bmatrix} : a, b, c \in \mathbb{R} \right\}$$

Note: $\begin{bmatrix} a & b \\ b & c \end{bmatrix} + \begin{bmatrix} x & y \\ y & z \end{bmatrix} = \begin{bmatrix} a+x & b+y \\ b+y & c+z \end{bmatrix}$

$k \begin{bmatrix} a & b \\ b & c \end{bmatrix} = \begin{bmatrix} ka & kb \\ kb & kc \end{bmatrix}$, so $\text{Sym}_2(\mathbb{R}) \subseteq M_{2 \times 2}(\mathbb{R})$.

Prop: Suppose A, B are $m \times n$ matrices and k is a scalar.

$$(A + kB)^T = A^T + k B^T$$

pf: $(A + kB)^T = ([a_{ij}] + k[b_{ij}])^T$

$$= [a_{ij} + kb_{ij}]^T$$

$$= [a_{ji} + kb_{ji}]$$

$$= [a_{ji}] + k[b_{ji}]$$

$$= [a_{ij}]^T + k[b_{ij}]^T = A^T + k B^T$$

□

M^T is obtained from M by swapping roles of rows and columns.

$$[m_{ij}]^T \neq [m_{ji}]$$

Cor: If A, B are symmetric and k is a scalar, then $A + kB$ is symmetric.

Pf: $(A + kB)^T = A^T + kB^T = A + kB$ □

Cor: The set of symmetric matrices is a subspace of the space of square matrices for every n .

(i.e. $\text{Sym}_n(\mathbb{R}) \leq M_{n \times n}(\mathbb{R})$).

Q: What is a nice basis of $\text{Sym}_n(\mathbb{R})$ (or $\text{Sym}_n(\mathbb{F})$)?

A: For $n=2$: $\begin{bmatrix} a & b \\ b & c \end{bmatrix} = \underline{a} \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} + \underline{b} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} + \underline{c} \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$

so $\left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\}$ span $\text{Sym}_n(\mathbb{F})$.

Lin. ind. follows because $kM_{i,j}$ has zeroes everywhere except (i,j) and (j,i) entries...

So $E_2 = \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\}$ is a basis.

For $n=3$: $\begin{bmatrix} a & b & c \\ b & d & e \\ c & e & f \end{bmatrix} = a \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} + b \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} + c \begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 0 \end{bmatrix}$
 $+ d \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} + e \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} + f \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$

$E_3 = \{ M_{i,j} : 1 \leq i \leq j \leq 3 \}$ is a basis of $\text{Sym}_3(\mathbb{F})$.

In general: $E_n = \{ M_{i,j} : 1 \leq i \leq j \leq n \}$ is a basis of $\text{Sym}_n(\mathbb{F})$

where $M_{i,j}$ has 1's in (i,j) and (j,i) , and 0's everywhere else.

Cor: $\dim(\text{Sym}_n(\mathbb{R})) = \frac{n(n-1)}{2} + n = \frac{n(n+1)}{2}$. □

Q: Is the product of symmetric matrices also symmetric?

Prop: Suppose A is an $(m \times k)$ -matrix and B is a $(k \times n)$ matrix.

Then $(AB)^T = B^T A^T$.

Pf: On hold. \square

Special Case: if $m = k = n = 2$:

$$(AB)^{-1} = B^{-1} A^{-1}$$

$$\left(\begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} x & y \\ z & w \end{bmatrix} \right)^T = \begin{bmatrix} ax + bz & ay + bw \\ cx + dz & cy + dw \end{bmatrix}^T$$
$$= \begin{bmatrix} ax + bz & cx + dz \\ ay + bw & cy + dw \end{bmatrix}$$

!!
☺

$$\begin{bmatrix} x & y \\ z & w \end{bmatrix}^T \begin{bmatrix} a & b \\ c & d \end{bmatrix}^T = \begin{bmatrix} x & z \\ y & w \end{bmatrix} \begin{bmatrix} a & c \\ b & d \end{bmatrix}$$
$$= \begin{bmatrix} xa + zb & xc + zd \\ ya + wb & yc + wd \end{bmatrix}$$

So if A and B are symmetric:

$$(AB)^T = B^T A^T = BA \neq AB$$

↑ Not always true !!

Ex: $A = \begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix}$, $B = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}$

Both ARE symmetric.

$$AB = \begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}$$

↑ NOT symmetric, \square

$(AB)^T = B^T A^T$ * always true

Prop: If A is invertible, then $(A^{-1})^T = (A^T)^{-1}$

Pf: $(A^{-1})^T A^T = (AA^{-1})^T = I^T = I \quad \therefore (A^T)^{-1} = (A^{-1})^T \quad \square$

Bad News: Products of symmetric matrices aren't symmetric $\ddot{}$.

Good News: We can still build symmetric matrices via product...

Consider any square matrix A .
 $(A^T A)^T = A^T (A^T)^T = A^T A$
So $A^T A$ is always symmetric. $\ddot{}$

Ex: $A = \begin{bmatrix} 2 & 4 \\ 1 & 3 \end{bmatrix}$. $A^T A = \begin{bmatrix} 2 & 1 \\ 4 & 3 \end{bmatrix} \begin{bmatrix} 2 & 4 \\ 1 & 3 \end{bmatrix} = \begin{bmatrix} 5 & 11 \\ 11 & 25 \end{bmatrix}$ \square

Q: What can the eigenvalues of a symmetric matrix be?

A (Forthcoming): If A is a real symmetric matrix,
the the eigenvalues of A are all real! \square

→ to give the full answer, we need to study more about
the complex vector spaces...

Defn: Let $z = a + bi$ be a complex number (w/ $a, b \in \mathbb{R}$).

The complex conjugate of z is $\bar{z} = \overline{a+bi} = a - bi$.

Ex: $\overline{3-i} = 3+i$, $\overline{5+7i} = 5-7i$, $\overline{\pi i} = -\pi i$, $\bar{e} = e$

Lemma: $\bar{z} = z$ if and only if $z \in \mathbb{R}$.

Pf: (\Rightarrow): If $\overline{a+bi} = a+bi$, then $a-bi = \overline{a+bi} = a+bi$,
so $2bi = 0$ yields $b = 0$.

(\Leftarrow): $\bar{a} = \overline{a+0i} = a-0i = a$ \square

NB: If $A \in M_{n \times n}(\mathbb{C})$, we can write $A = \operatorname{Re}(A) + i \operatorname{Im}(A)$
where both $\operatorname{Re}(A)$ and $\operatorname{Im}(A)$ are real matrices.

$$\underline{\text{Ex:}} A = \begin{bmatrix} 1+i & 1-i \\ 3+2i & 5-i \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 3 & 5 \end{bmatrix} + \begin{bmatrix} i & -i \\ 2i & -i \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 3 & 5 \end{bmatrix} + i \begin{bmatrix} 1 & -1 \\ 2 & -1 \end{bmatrix}.$$

$$\operatorname{Re}(A) = \begin{bmatrix} 1 & 1 \\ 3 & 5 \end{bmatrix}, \quad \operatorname{Im}(A) = \begin{bmatrix} 1 & -1 \\ 2 & -1 \end{bmatrix}.$$

Point: we can extend the definition of conjugate to matrices!

$$\overline{A} = \overline{\operatorname{Re}(A) + i \operatorname{Im}(A)} = \operatorname{Re}(A) - i \operatorname{Im}(A)$$